

Problem 1

Proof. Let $K = \{k + 1, \dots, 2k + 1\}$.

Part A

Assume towards contradiction that A is not sum free. Then there exists some b_i, b_j, b_k such that $b_i + b_j = b_k$. Since $b_i, b_j, b_k < \max \{|b_i| : 1 \leq i \leq n\}$ and $p > 2 \max \{|b_i| : 1 \leq i \leq n\}$, $b_i + b_j < p$ thus

$$b_i + b_j \equiv b_k \pmod{p} \implies xb_i + xb_j \equiv xb_k \pmod{p}.$$

But this means that $d_i + d_j = d_k$, which are elements of a sum free subset. This is a contradiction, hence A must be sum free.

Part B

Let D_i be the indicator random variable that $d_i \in K$. Then we have

$$\mathbb{E}[|\{d_1, \dots, d_n\} \cap K|] = \mathbb{E}\left[\sum_{i=1}^n D_i\right] = \sum_{i=1}^n \mathbb{E}[D_i].$$

Since $x \neq 0$ and $b_i \neq 0$, $d_i \neq 0$ and hence the number of total possibilities for d_i is $p - 1$. Therefore since x is chosen uniformly at random

$$\mathbb{P}[d_i \in K] = \frac{(2k + 2) - (k + 1)}{p - 1} = \frac{k + 1}{3k + 1}$$

Thus we have that the desired expectation equals

$$\sum_{i=1}^n \mathbb{E}[D_i] = \sum_{i=1}^n \frac{k + 1}{3k + 1} = n \left(\frac{k + 1}{3k + 1} \right).$$

Since $\frac{k+1}{3k+1} > \frac{1}{3}$ for all $k \geq 0$,

$$\mathbb{E}[|\{d_1, \dots, d_n\} \cap K|] > \frac{n}{3}.$$

Part C

Since the expectation is greater than $\frac{n}{3}$, it follows that

$$\mathbb{P}\left[|\{d_1, \dots, d_n\} \cap K| > \frac{n}{3}\right] > 0.$$

Therefore there must exist some x such that

$$|\{d_1, \dots, d_n\} \cap K| > \frac{n}{3}.$$

Part D

Take A to be the collection of associated b_i to x such that $d_i \in K$. Then by (a) this set is A sum free and by (c) this set has size greater than $\frac{n}{3}$. \diamond

Problem 2

Proof. Let $\varepsilon_1, \dots, \varepsilon_n$ be independently randomly chosen weights such that $\mathbb{P}[\varepsilon_i = -1] = \mathbb{P}[\varepsilon_i = 1] = \frac{1}{2}$. If $\langle \cdot, \cdot \rangle$ is the Euclidean inner product, then $\|\cdot\|^2 = \langle \cdot, \cdot \rangle$ and thus by linearity of the inner product

$$\begin{aligned} \left\| \sum_i \varepsilon_i v_i \right\|^2 &= \left\langle \left(\sum_i \varepsilon_i v_i \right), \left(\sum_j \varepsilon_j v_j \right) \right\rangle \\ &= \sum_i \left\langle \varepsilon_i v_i, \sum_j \varepsilon_j v_j \right\rangle \\ &= \sum_i \sum_j \langle \varepsilon_i v_i, \varepsilon_j v_j \rangle \\ &= \sum_i \sum_j \varepsilon_i \varepsilon_j \langle v_i, v_j \rangle. \end{aligned}$$

Note that since $\mathbb{E}[\varepsilon_i^2] = 1$ and for $i \neq j$, $\mathbb{E}[\varepsilon_i \varepsilon_j] = \mathbb{E}[\varepsilon_i] \mathbb{E}[\varepsilon_j] = 0$,

$$\mathbb{E}[\varepsilon_i \varepsilon_j] = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}.$$

Therefore by linearity of expectation, we have

$$\begin{aligned}
 \mathbb{E} \left[\left\| \sum_i \varepsilon_i v_i \right\|^2 \right] &= \mathbb{E} \left[\sum_i \sum_j \varepsilon_i \varepsilon_j \langle v_i, v_j \rangle \right] \\
 &= \sum_i \sum_j \mathbb{E}[\varepsilon_i \varepsilon_j] \langle v_i, v_j \rangle \\
 &= \sum_i \langle v_i, v_i \rangle \\
 &= n
 \end{aligned}$$

since all of the v_i are unit vectors. Therefore since the expected value of the norm squared is n , then the probability that there exists an assignment of the ε_i that makes the norm squared less than or equal to n is non zero. Since this probability is non zero, there then must exist such an assignment. Thus

$$\left\| \sum_i \varepsilon_i v_i \right\|^2 \leq n \implies \left\| \sum_i \varepsilon_i v_i \right\| \leq \sqrt{n}$$

which was to be shown. \diamond

Problem 3

Proof. Trivially, if $n < 3$ then K_n has no possible rainbow triangles and $\binom{n}{3} \frac{2}{9} < 1$. Assume then that $n \geq 3$. Consider a random assignment of a color to each edge $e \in K_n$ such that each color is equally likely. Let $X_{i,j,k}$ be the indicator random variable that the triangle comprised of edges i, j, k is rainbow. Then the expected number of rainbow triangles is

$$\mathbb{E} \left[\sum_{i \neq j \neq k} X_{i,j,k} \right] = \sum_{i \neq j \neq k} \mathbb{E}[X_{i,j,k}] = \binom{n}{3} \mathbb{E}[X_{1,2,3}]$$

since each $X_{i,j,k}$ is identically distributed hence we only need to find the expectation of a single one, and there are $\binom{n}{3}$ edges since K_n is complete.

The expectation is simply the probability a given triangle is rainbow. There are $3^3 = 27$ possible colorings of the edges of a triangle, and there are $3 \cdot 2 \cdot 1 = 6$ colorings in which all edges are different colors. Since every coloring is equally likely, this means

$$\mathbb{E} \left[\sum_{i \neq j \neq k} X_{i,j,k} \right] = \binom{n}{3} \mathbb{E}[X_{1,2,3}] = \binom{n}{3} \frac{6}{27} = \binom{n}{3} \frac{2}{9}.$$

Therefore the probability that a random edge coloring of K_n contains more than $\binom{n}{3} \frac{2}{9}$ rainbow triangles is non zero, hence such a random edge coloring must exist. \diamond

Problem 4

Proof. Let $S = \{1, \dots, n\}$ and $I = (c, c+1)$ where $c \in \mathbb{R}$. Given a vector $\varepsilon \in \{0, 1\}^n$, we can associate a subset $E \subseteq S$ in which $i \in E$ iff $\varepsilon_i = 1$. Note then that the sum

$$\sum_{i=1}^n \varepsilon_i a_i = \sum_{i \in E} a_i.$$

Assume towards contradiction we have a family of such subsets such that they satisfy the unit interval summation property, but there is two subsets E_1 and E_2 such that $E_1 \neq E_2$ and $E_1 \subset E_2$. Note that

$$c < \sum_{i \in E_1} a_i < c + 1.$$

Since $E_1 \subset E_2$,

$$\sum_{i \in E_2} a_i = \sum_{i \in E_1} a_i + \sum_{i \in E_2 \setminus E_1} a_i.$$

Furthermore, there must be at least one element a_i in E_2 that isn't in E_1 . Combined with the fact that all $a_i \geq 1$,

$$\sum_{i \in E_2 \setminus E_1} a_i \geq 1.$$

Thus

$$\sum_{i \in E_2} a_i = \sum_{i \in E_1} a_i + \sum_{i \in E_2 \setminus E_1} a_i \geq 1 + \sum_{i \in E_1} a_i > c + 1.$$

But this means E_2 can not be in the family since it does not lay in I , a contradiction. Therefore the family of E_i is a Sperner Family and hence has at most $\binom{n}{\frac{n}{2}} = \binom{n}{\lfloor \frac{n}{2} \rfloor}$ members. By the initial equivalence, this means that a family of ε vectors is also subject to the same upper bound. \diamond

Problem 5

Proof. Let $\tilde{X} = X - a \cdot \mathbb{1}_{X \geq a}$. Note that $\tilde{X} \geq 0$, since when $X < a$ then $\tilde{X} = X$ and when $X \geq a$, then $\tilde{X} = X - a \geq a - a = 0$. Furthermore, by linearity of expectation $\mathbb{E}[\tilde{X}] = \mathbb{E}[X] - a\mathbb{P}[X \geq a]$. When $\mathbb{E}[\tilde{X}] = 0$, we get equality of the Markov inequality. Since \tilde{X} is non-negative, the expectation being 0 means that $\mathbb{P}[\tilde{X} = 0] = 1$. The only two possibilities for X that satisfy this are either $\mathbb{P}[X = 0] = 1$ or $\mathbb{P}[X = a] = 1$. \diamond